

1. Introduction:

- \succ In image segmentation and visual tracking, well annotated image and video ground truth are essential for performance evaluation and comparison of methods.
- Image annotation based on bounding boxes, coarse polygons and region masks (obtained from graph cuts or level set) are either inaccurate or less flexible.
- > How to acquire accurate image annotations with light human workload still remains an open problem.
- \succ In this paper, we develop a boundary based method for accurate image annotating. This tool outperforms other similar polygon-based methods by solving following problems.

(1). Sampling problem: Many control points are required to describe smoothed and complex curves.



- \checkmark To describe the given circle as accurately as possible, many control points have to be selected.
- \checkmark That would be a very heavy work when annotating similar or even complex targets.

(2). Locating problem: Accurately locating these control points are difficult.



- \checkmark Red, green, blue and yellow pixels are possible control points.
- ✓ Users have to make a decision of picking which one as the control point based on their vision and intuition.
- \checkmark Many times of decision making are excruciating.

(3). Description problem: It is hard to describe objects with holes and objects divided by occlusions.



Object with hole

Object divided by occlusions



ByLabel: A Boundary Based Semi-Automatic Image Annotation Tool Xuebin Qin, Shida He, Zichen Zhang, Masood Dehghan and Martin Jagersand Department of Computing Science, University of Alberta



(d) Output region mask Fig. 5. Annotate an object step by step

Code: http://webdocs.cs.ualberta.ca/~vis/bylabel/

3. Results:

We evaluated our method by asking ten volunteers to annotate 15 typical images using our tool **ByLabel** and LabelMe. The results of clicks, time and error are shown as in Tab. 1, Tab. 2 and Tab. 3 respectively.

Iable. 1. Average Clicks																
Image	A	A1 A	2 A	.3 A	4 A5	5 B1	B2	B3	B4	B5	C1	C 2	C 3	C 4	C5	Average
LabelMe	e :	5 9	9 1	3 3	5 123	3 37	28	45	38	51	98	73	129	147	177	67
Ours	-	5 3	5 4	4 3	3 21	13	2	11	23	11	35	16	29	37	62	18
Table. 2. Average Time Costs (s)																
Image	A1	A2	A3	A4	A5	B 1	B2	B 3	B 4	B5	C 1	C2	C3	C4	C5	Average
LabelMe	9.61	14.80	18.80	37.85	113.71	44.13	32.08	52.63	49.62	47.80	90.71	74.56	104.79	145.52	144.39	65.04

Iable. 1. Average Clicks																
Image	А	.1 A	.2 A	.3 A	4 A5	5 B1	B2	B3	B 4	B5	C 1	C 2	C 3	C 4	C5	Average
LabelMe		5 9	9 1	3 3	5 123	3 37	28	45	38	51	98	73	129	147	177	67
Ours	4	5 5	5 4	4 3	3 21	13	2	11	23	11	35	16	29	37	62	18
Table. 2. Average Time Costs (s)																
Image	A1	A2	A3	A4	A5	B 1	B2	B 3	B 4	B5	C 1	C2	C 3	C 4	C5	Average
LabelMe	9.61	14.80	18.80	37.85	113.71	44.13	32.08	52.63	49.62	47.80	90.71	74.56	104.79	145.52	144.39	65.04
Ours	4.51	4.26	4.01	2.56	23.25	20.03	2.60	13.49	28.19	12.26	64.30	31.02	47.07	62.76	107.62	28.53

Table. 3. Average Error (pixel)																
Image	A1	A2	A3	A4	A5	B1	B2	B 3	B4	B5	C1	C2	C 3	C 4	C5	Average
LabelMe Ours	1.23 0.50	1.35 0.74	1.50 0.86	1.35 0.64	1.08 0.67	1.21 0.14	1.09 0.00	0.78 0.01	0.73 0.04	0.91 0.02	0.97 0.16	1.25 0.22	1.05 0.09	0.92 0.03	1.02 0.37	1.10 0.30

Other typical annotation results are shown as follows:

(a) Pedestrian

(d) Input image

(e) Classes annotation (f) Instances annotation Fig. 6. Typical annotation results

4. Conclusions:

- \succ We develop a novel semi-automatic boundary based image annotation tool, **ByLabel**.
- > ByLabel introduces edge detection and splitting algorithms to assist
- > Additionally, **ByLabel** can also be used to annotate video streams frame by frame.

(c) Bicycle

annotation, which greatly improves the annotation efficiency and accuracy.